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ON THE RATE OF SUPERLINEAR CONVERGENCE OF A CLASS OF VARIABLE M--ETC(U)

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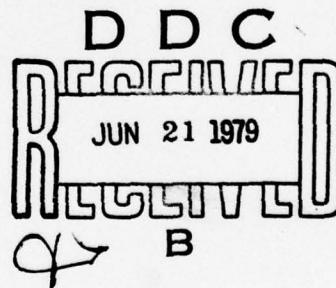
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ON THE RATE OF SUPERLINEAR CONVERGENCE  
OF A CLASS OF VARIABLE METRIC METHODS

Klaus Ritter

Dedicated to Professor Dr. H. Görtler  
on the occasion of his seventieth birthday

Technical Summary Report #1950

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ABSTRACT

This paper considers a class of variable metric methods for unconstrained minimization. Without requiring exact line searches each algorithm in this class converges globally and superlinearly. Various results on the rate of the superlinear convergence are obtained.

AMS (MOS) Subject Classification: 90C30

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## SIGNIFICANCE AND EXPLANATION

Many practical problems in operations research may be reduced to minimizing a function without constraints. In this paper we discuss a class of algorithms for unconstrained minimization problems which converge to the solution from an arbitrary starting point. In order to judge the efficiency of such an algorithm estimates for the rate of convergence are important. Such estimates are derived in this paper.

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ON THE RATE OF SUPERLINEAR CONVERGENCE  
OF A CLASS OF VARIABLE METRIC METHODS

Klaus Ritter

1. Introduction

Variable metric methods have been used successfully in unconstrained minimization. Under appropriate assumptions such a method generates a sequence  $\{x_j\}$  which converges superlinearly to a global minimum. It is the purpose of this paper to study the rate of the superlinear convergence.

A first result concerning the rate of superlinear convergence of a particular variable metric method, the Davidon-Fletcher-Powell method [4], [6], was obtained by Burmeister [3]. Assuming that the optimal step size is used he proved that this method generates a sequence which converges  $n$ -step quadratically when applied to a function  $F(x)$  depending on  $n$  variables. Using a non-optimal step size Stoer [14] showed that this result is valid for a class of variable metric methods, the so-called restricted Broyden-methods [1], provided the initial point is sufficiently close to a minimizer of  $F(x)$ . Assuming that for every iteration the last  $n$  search directions are uniformly linearly independent and using an appropriate non-optimal step size Schuller [13] proved that the sequence  $\{\|x_{j+1}-z\|/\|x_j-z\| \mid \|x_{j-n}-z\| \text{ is bounded, where } z \text{ is a minimizer of } F(x) \text{ and the sequence } \{x_j\} \text{ is generated by the Broyden-Fletcher-Goldfarb-Shanno - method [2], [7], [8], [12].}\}$

In this paper we will be concerned with the restricted Broyden methods. These methods have the property that they maintain the symmetry and positive-definiteness of the matrix used to approximate the Hessian matrix of  $F(x)$ . They form a subclass of the Huang class [9] of variable metric methods. Throughout the paper a non-optimal step size, based on a quadratic interpolation formula, is used.

First we will generalize Schuller's result by extending it to all restricted Broyden methods. Secondly we will strengthen Stoer's result by removing the assumption that the initial point has to be close to a minimizer of  $F(x)$  and by showing that the sequence  $\{\|x_{j+n}-z\|/\|x_j-z\|\}^2$  is not only bounded but converges to zero. Finally assuming that a certain lower bound on the rate of convergence is valid we will show that the sequence  $\{\|x_{j+1}-z\|/\|x_j-z\| \mid \|x_{j-n+1}-z\| \text{ is bounded and that the search directions are asymptotically conjugate with respect to the Hessian matrix of } F(x) \text{ at the global minimizer.}\}$



## 2. A class of variable metric methods

Let  $x \in \mathbb{R}^n$  and let  $F(x)$  be a real-valued function. If  $F(x)$  is twice differentiable at a point  $x_j$  we denote the gradient and the Hessian matrix of  $F(x)$  at  $x_j$  by  $g_j = \nabla F(x_j)$  and  $G_j = G(x_j)$ , respectively. A prime is used for the transpose of a vector or a matrix. For any  $x \in \mathbb{R}^n$ ,  $\|x\|$  denotes the Euclidean norm of  $x$ .

We consider the problem of determining a sequence

$$(2.1) \quad x_{j+1} = x_j - \sigma_j g_j, \quad j = 0, 1, 2, \dots$$

which converges to a global minimizer  $x$ , say, of  $F(x)$ . Here the vector  $g_j$  is called a search direction and the scalar  $\sigma_j$  is referred to as the step size.

If a variable metric method is used to compute the sequence (2.1) then an

$(n, n)$ -matrix  $H_j$  is associated with each  $x_j$  and

$$(2.2) \quad \sigma_j = H_j g_j.$$

The matrix  $H_{j+1}$  is determined from  $H_j$  by adding a rank one or two matrix in such a way that  $H_{j+1}$  satisfies the quasi-Newton equation

$$(2.3) \quad H_{j+1} d_j = p_j,$$

where

$$d_j = \frac{g_j - g_{j+1}}{\|g_j\|}, \quad p_j = \frac{s_j}{\|s_j\|}.$$

The various variable metric methods differ in the update procedure which is used to compute  $H_{j+1}$  from  $H_j$ . A large class of such methods has been studied by Broyden [1], Fletcher [7], Huang [9], and Dixon [5]. In the following we will consider a subclass of these update procedures which ensure that if the initial matrix  $H_0$  is symmetric all matrices  $H_j$  are symmetric. It has been shown in

(11) that the update formulas that correspond to this subclass can be written in the form

$$(2.4) \quad H_{j+1} = H_j + \frac{\beta_1 (d_j p_j - d_j' H_j d_j) + \beta_2 d_j' H_j d_j}{d_j' p_j (\beta_1 d_j p_j + \beta_2 d_j' H_j d_j)} p_j p_j' - \frac{\beta_1 \frac{p_j d_j' H_j d_j p_j}{\beta_1 d_j p_j + \beta_2 d_j' H_j d_j} - \beta_2 \frac{H_j d_j d_j' H_j}{\beta_1 d_j p_j + \beta_2 d_j' H_j d_j}}{d_j' p_j},$$

where  $\beta_1$  and  $\beta_2$  are arbitrary parameters with  $\beta_1^2 + \beta_2^2 > 0$ .

Choosing  $\beta_1 = 1$ ,  $\beta_2 = 0$  and  $\beta_1 = 0$ ,  $\beta_2 = 1$  we obtain the two special cases

$$H_{j+1} = H_j + \frac{d_j p_j - d_j' H_j d_j}{(d_j' p_j)^2} p_j p_j' - \frac{p_j d_j' H_j d_j p_j}{d_j' p_j}$$

and

$$H_{j+1} = H_j + \frac{p_j p_j'}{d_j' p_j} - \frac{H_j d_j d_j' H_j}{d_j' p_j}$$

which are known as BFGS - method (Broyden [2]), Fletcher [7], Goldfarb [8], Shanno [12]) and DFP - method (Davidon [4], Fletcher, Powell [6]), respectively.

Assuming that  $H_0$  is positive definite and that, for all  $j$ ,

$$g_{j+1}' p_j < g_j' p_j, \quad \text{i.e., } d_j' p_j = \frac{g_j' p_j - g_{j+1}' p_j}{\|g_j\|} > 0$$

we conclude from Lemma 1 in [11] that

$$\beta_1 \beta_2 \geq 0, \quad \beta_1 + \beta_2 \neq 0$$

is a sufficient condition for all matrices  $H_j$  to be positive definite.

If  $H_j$  is positive definite it has been shown in [11] that  $H_j$  can be written in the form

$$(2.5) \quad H_j = \frac{p_j p_j'}{g_j' p_j} + \sum_{i=1}^j \frac{g_i g_i'}{g_i' p_i} + \sum_{i=1}^j \frac{p_i p_i'}{d_i' p_i}.$$

where

$$i) \quad p_j = \frac{H_j q_j}{\|H_j q_j\|}, \quad p_j = \frac{1}{\|H_j q_j\|},$$

$$ii) \quad w_j \in \text{span}(q_j, q_{j+1}) \text{ such that } w_j^T p_j = 0 \text{ and } q_j = H_j w_j$$

has norm one,

$$iii) \quad \text{the vectors } d_1, \dots, d_n \text{ are orthogonal to } p_j \text{ and } q_j \text{ and are such that}$$

$$d_i^T H_j d_k = 0, \quad i, k = 1, \dots, n, \quad i \neq k$$

and

$$F_{ij} = H_j d_i, \quad i = 1, \dots, n, \text{ has norm one.}$$

Then every  $H_{j+1}$  determined by (2.4) has the form (see (11)),

$$(2.6) \quad H_{j+1} = \frac{p_j p_j^T}{d_j^T p_j} + w_j w_j^T + \sum_{i=3}^n \frac{p_i p_i^T}{d_i^T p_j},$$

where the vector  $w_j$  is uniquely determined by the conditions

$$w_j \in \text{span}(q_j, p_j), \quad \|w_j\| = 1, \quad d_j^T w_j = 0, \quad w_j^T u_j > 0$$

and the parameter  $w_j$  depends on the particular numbers  $\beta_1$  and  $\beta_2$  used in (2.4). More precisely,

$$(2.7) \quad w_j = \gamma_j \|q_j - \frac{d_j^T q_j}{d_j^T p_j} p_j\|$$

with

$$\gamma_j = \frac{\beta_1 d_j^T p_j + \beta_2 \frac{(d_j^T p_j)^2}{d_j^T p_j}}{\beta_1 d_j^T p_j + \beta_2 \frac{(d_j^T p_j)^2}{d_j^T p_j}}.$$

We now assume that  $s_j$  is determined by (2.2) and  $H_j$  is determined by the update formula (2.4) with  $\beta_1 + \beta_2 \neq 0$ . Under appropriate assumptions on  $F(x)$  and on the choice of the step size  $\sigma_j$  it has been shown in [12] that the sequence (2.1) converges globally and superlinearly to a global minimizer of  $F(x)$ .

In particular if  $F(x)$  is convex and twice continuously differentiable and the sequence  $\{x_j\}$  converges to some  $z$ , say, such that  $\nabla F(z) = 0$ ,  $G = G(z)$  is positive definite and the Lipschitz condition

$$(2.8) \quad \|G(x) - G(z)\| \leq L \|x - z\|$$

is satisfied for all  $x$  in some neighborhood of  $z$ , then it follows from Theorem 3 in [11] that

$$i) \quad \frac{\|x_{j+1} - z\|}{\|x_j - z\|} \rightarrow 0 \text{ as } j \rightarrow \infty.$$

$$ii) \quad \text{The sequences } \{H_j\} \text{ and } \{H_j^{-1}\} \text{ are bounded.}$$

$$iii) \quad \sigma_j = \sigma_j^*$$

$$(2.9) \quad \sigma_j^* = \frac{q_j^T q_j}{2(F(x_j - s_j) - F(x_j) + q_j^T s_j)}$$

is an acceptable step size for  $j$  sufficiently large, provided  $\beta_1 \beta_2 \geq 0$ .

It is not difficult to verify that if  $\sigma_j^*$  is defined and positive then it is the global minimizer of the quadratic function  $Q_j(\alpha)$  which has the properties

$$Q_j(0) = F(x_j), \quad Q_j(1) = F(x_j - s_j), \quad \frac{d}{d\alpha} Q(\alpha) = \frac{d}{d\sigma} F(x_j - \sigma s_j) \text{ for } \alpha = 0.$$

The above assumptions imply that there are constants  $0 < \mu < \eta$  such that for every  $x$  in some neighborhood of  $z$ ,

$$(2.10) \quad \mu \|y\|^2 \leq y^T G(x) y \leq \eta \|y\|^2 \text{ for all } y \in \mathbb{R}^n.$$

Deleting finitely many members of the sequence  $\{x_j\}$ , if necessary, we may therefore assume that  $\sigma_j = o_j^2$  for all  $j$  and that there is some neighborhood  $U(z)$  of  $z$  such that  $\{x_j\} \subset U(z)$  and the inequalities (2.8) and (2.10) hold for every  $x \in U(z)$ .

In view of these results we are justified in requiring that the following assumption is satisfied.

#### Assumption 1

- i) The sequence  $\{H_j\}$  is determined by (2.4) with  $\beta_1, \beta_2 \geq 0$ ,  $\beta_1 + \beta_2 \neq 0$  and  $H_0$  symmetric and positive definite.
- ii) The sequences  $\{s_j\}$ ,  $\{o_j\}$  and  $\{x_j\}$  are determined by (2.2), (2.9), and (2.1), respectively.
- iii) There is some  $z$  and a neighborhood  $U(z)$  such that  $F(x)$  is twice continuously differentiable on  $U(z)$ ,  $\forall F(x) = 0$ ,  $\{x_j\} \subset U(z)$  and the inequalities (2.8) and (2.10) are satisfied for every  $x \in U(z)$ .

$$\text{iv)} \quad \frac{\|x_{j+1} - z\|}{\|x_j - z\|} \rightarrow 0 \text{ as } j \rightarrow \infty.$$

- v) There are numbers  $0 < \tau_1 < \tau_2$  such that  $\tau_1 \|x\|^2 \leq x' H_j x \leq \tau_2 \|x\|^2$  for all  $x \in E^n$  and  $j = 0, 1, \dots$ .

For later reference we state the following lemma.

#### Lemma 1

Let Assumption 1 be satisfied. Then the following statements hold.

- i)  $0 < \gamma_j < 1$  and  $\gamma_j \rightarrow 1$  as  $j \rightarrow \infty$ .
- ii)  $|\gamma_j - 1| = O((d_j' q_j)^2) = O\left(\frac{\|q_{j+1}\|^2}{\|q_j\|^2}\right)$ .
- iii)  $|q_{j+1}' p_j| = O(\|q_j\|^2)$ ,  $\sigma_j \rightarrow 1$  as  $j \rightarrow \infty$ .

$$\text{iv)} \quad \|d_j - \sigma_j p_j\| = O(\|q_j\|), \quad d_j' p_j \geq \nu, \quad \|d_j\| \leq \eta.$$

$$\text{v)} \quad \|x_j - z\| = O(\|q_j\|), \quad \|q_j\| = O(\|x_j - z\|).$$

The first three statements of the lemma have been proved in [1]. The last two parts of the lemma are well-known consequences of part iii) of Assumption 1. They can be proved by a simple application of Taylor's theorem.

#### Remark

It follows from part ii) of Lemma 1 that

$$\frac{\gamma_j - 1}{d_j' q_j} \rightarrow 0 \text{ as } d_j' q_j \rightarrow 0.$$

Therefore, we define  $(\gamma_j - 1)/d_j' q_j$  to be zero if  $d_j' q_j = 0$ .



### 2. Rates of superlinear convergence

Throughout this section we require that Assumption 1 is satisfied.

By (2.3) we have for every  $j$

$$H_{j+1}d_j = P_j \cdot d_j = \frac{q_j^{-q_{j+1}}}{\|q_j\|}, P_j = \frac{s_j}{\|s_j\|}.$$

However, if  $i < j - 1$  then in general

$$H_j d_i \neq P_i$$

and one of the main difficulties in estimating the rate of superlinear convergence of the sequence  $\{x_j\}$  is to find upper bounds for the numbers

$$\|H_j d_i - P_i\|, i = j-2, \dots, j-n.$$

As a first step towards this goal we prove the following two lemmas.

#### Lemma 2

For every  $j$  and every  $i < j$  let

$$M_j = I - \frac{P_j d_i}{d_j^T P_j} \quad \text{and} \quad v_{ij} = H_j d_i - P_i.$$

Then

$$1) \quad H_j d_j = \frac{1}{\gamma_j} \left[ P_j \frac{q_{j+1}^T P_j}{q_j^T P_j} - P_{j+1} \frac{\|s_{j+1}\|}{q_j^T P_j} d_j^T P_j \right]$$

$$2) \quad v_{i,j+1} = M_j v_{ij} + P_j \frac{P_j d_i - d_j^T P_j}{d_j^T P_j} + \left( \frac{\gamma_j - 1}{\gamma_j} \frac{d_j^T q_j}{d_j^T P_j} - \frac{P_j d_i}{d_j^T P_j} \right) \left( P_j \frac{q_{j+1}^T P_j}{q_j^T P_j} - P_{j+1} \frac{\|s_{j+1}\|}{q_j^T P_j} d_j^T P_j \right).$$

Proof:

1) Observing that by (2.5)

$$(3.1) \quad H_j d_j = P_j \frac{d_j^T P_j}{d_j^T P_j} + q_j \frac{d_j^T q_j}{w_j^T q_j}$$

we obtain

$$(3.2) \quad \begin{aligned} H_j d_j &= P_j \frac{d_j^T P_j}{d_j^T P_j} + q_j \frac{d_j^T q_j}{w_j^T q_j} - \frac{P_j}{d_j^T P_j} \left( \frac{(d_j^T P_j)^2}{d_j^T P_j} + \frac{(d_j^T q_j)^2}{w_j^T q_j} \right) \\ &= d_j^T q_j \frac{q_{j+1}^T P_j}{w_j^T q_j}, \quad q_j = -\frac{d_j^T q_j}{d_j^T P_j}. \end{aligned}$$

Furthermore, because

$$w_j^T P_j = d_j^T u_j = 0, \quad q_{j+1} = q_j - \|s_j\| d_j$$

and by (2.16) in [11]

$$u_j = \frac{q_{j+1}^T P_j}{\|q_{j+1}^T P_j\|}$$

it follows from (2.6), (2.7), and (3.2) that

$$\begin{aligned} s_{j+1} &= H_{j+1} q_{j+1} = P_j \frac{P_j q_{j+1}}{d_j^T P_j} + u_j \frac{w_j^T q_{j+1}}{w_j^T u_j} \\ &= P_j \frac{P_j q_{j+1}}{d_j^T P_j} + (q_{j+1}^T P_j) \frac{\gamma_j d_j^T P_j}{w_j^T q_j} \\ &= P_j \frac{P_j q_{j+1}}{d_j^T P_j} - \gamma_j \frac{P_j q_j}{d_j^T P_j} d_j^T q_j \frac{(q_{j+1}^T P_j)}{w_j^T q_j} \\ &= P_j \frac{P_j q_{j+1}}{d_j^T P_j} - \gamma_j \frac{P_j q_j}{d_j^T P_j} H_j d_j. \end{aligned}$$

Therefore,

$$H_j d_j = \frac{1}{\gamma_j} \left[ P_j \frac{P_j q_{j+1}}{d_j^T P_j} - P_{j+1} \frac{\|s_{j+1}\|}{d_j^T P_j} d_j^T P_j \right].$$

11) Because by (3.1)

$$\frac{P_j d_i}{d_j^T P_j} = \frac{P_j P_j}{d_j^T P_j} - q_j \frac{P_j q_j}{w_j^T q_j}$$

it follows from the definition of  $H_{j+1}$  and (3.2) that

$$\begin{aligned} H_{j+1} &= H_j - \frac{P_j P_j^*}{P_j^* P_j} + \frac{P_j P_j^*}{P_j^* P_j} - \frac{q_j q_j^*}{w_j^* q_j} + \gamma_j \frac{(q_j^* a_j P_j)(q_j a_j P_j^*)}{w_j^* q_j} \\ &= H_j - \frac{P_j d_j^* H_j}{d_j^* P_j} - \frac{(q_j^* a_j P_j) q_j^*}{w_j^* q_j} + \frac{P_j^* P_j^*}{d_j^* P_j} + \gamma_j \frac{(q_j^* a_j P_j)(q_j a_j P_j^*)}{w_j^* q_j} \\ &= M_j H_j + \frac{P_j P_j^*}{d_j^* P_j} + (\gamma_j - 1) \frac{(q_j^* a_j P_j) q_j^*}{w_j^* q_j} + \gamma_j \frac{(q_j^* a_j P_j) a_j P_j^*}{w_j^* q_j} \\ &= M_j H_j + \frac{P_j P_j^*}{d_j^* P_j} + d_j^* q_j \frac{(q_j^* a_j P_j)}{w_j^* q_j} \left[ \frac{\gamma_j - 1}{d_j^* q_j} q_j^* - \frac{\gamma_j}{d_j^* P_j} P_j^* \right] \\ &= M_j H_j + \frac{P_j P_j^*}{d_j^* P_j} + M_j H_j d_j \left( \frac{\gamma_j - 1}{d_j^* q_j} q_j^* - \frac{\gamma_j}{d_j^* P_j} P_j^* \right). \end{aligned}$$

Hence,

$$\begin{aligned} v_{j+1} &= H_{j+1} d_{j+1} - P_{j+1} \\ &= M_j v_{j+1} + P_j \frac{P_j^* d_j^* P_j}{d_j^* P_j} + M_j H_j d_j \left( \frac{\gamma_j - 1}{d_j^* q_j} q_j^* d_{j+1} - \frac{\gamma_j}{d_j^* P_j} P_{j+1} \right). \end{aligned}$$

In conjunction with part 1) this equality completes the proof of the lemma.

### Lemma 3

For every  $j \geq n$  and  $j-n \leq i < l \leq j$ ,

$$v_{il} = 0 \text{ if } l = i+1$$

$$\|v_{il} - \eta_{il} \frac{\|s_l\|}{g_{l-1}^* P_{l-1}} P_l\| = o(\|q_l\|) \text{ if } l > i+1,$$

where

$$\eta_{il} = \tau_{l-1}^* d_{l-1}^* q_{l-1}^* d_{l-1}^* P_{l-1} - P_{l-1}^* d_{l-1}^* \tau_{l-1} = \frac{\gamma_{l-1}^{-1}}{\gamma_{l-1}} \frac{1}{d_{l-1}^* q_{l-1}^*}$$

and

$$|\eta_{il}| = o(1), \quad |\tau_{l-1}| = o\left(\frac{\|q_l\|}{\|g_{l-1}\|}\right).$$

Proof.

Since every  $H_l$  satisfies the quasi-Newton equation (2.3) we have  $v_{l-1,l} = H_l d_{l-1} - P_{l-1} = 0$ . Let  $1 < k < l$ . By part 1) of Lemma 2

$$(3.3) \quad v_{l,k+1} = M_{l,k+1}^* v_{lk} - \eta_{l,k+1} \frac{P_{k+1}}{g_k^* P_k} + v_{l,k+1} P_k,$$

where

$$\eta_{l,k+1} = \tau_k d_k^* q_k^* d_k^* P_k - P_k^* d_k^* \tau_k = \frac{\gamma_k^{-1}}{\gamma_k} \frac{1}{d_k^* q_k^*}$$

and

$$v_{l,k+1} = \frac{P_k^* d_k^* - d_k^* P_k}{d_k^* P_k} + \left( \tau_k d_k^* q_k^* - \frac{P_k^* d_k^*}{d_k^* P_k} \right) \frac{g_{k+1}^* P_k}{g_k^* P_k}.$$

Since by Lemma 1,  $\|d_j\| \leq \eta_j, \gamma_j \rightarrow 1$  as  $j \rightarrow \infty$ ,

$$|1 - \gamma_j| = o(|d_j^* q_j|^2) \text{ and } |d_j^* q_j| = o\left(\frac{\|g_{j+1}\|}{\|g_j\|}\right)$$

we have

$$(3.4) \quad |\tau_k| = o\left(\frac{\|g_{k+1}\|}{\|g_k\|}\right) \text{ and } |\eta_{l,k+1}| = o(1).$$

Furthermore by Lemma 1,  $\|d_j - g_j\| = o(\|g_j\|)$  which implies

$$(3.5) \quad |P_k^* d_k - d_k^* P_k| = o(\|g_k\|).$$

Therefore,

$$(3.6) \quad |v_{l,k+1}| = o(\|g_l\|)$$

because it follows from Lemma 1 that

$$d_{k+1}^* P_k \geq \mu > 0 \text{ and } |q_{k+1}^* P_k|/q_k^* P_k = o(\|g_k\|).$$

Since  $\forall_{i=1,2,3}$  the sequence  $\{M_j\}$  is bounded and, for every  $j$ ,  $M_j P_j = 0$ , the statement of the lemma follows now from (3.4), (3.6) and the equality (3.3).

In order to obtain the first result on the rate of superlinear convergence from the above lemma we make the assumption that, for  $j$  sufficiently large,  $n$  consecutive search directions are uniformly linearly independent.

#### Assumption 2

For  $j$  sufficiently large  $P_j^{-1}$  exists and  $\{P_j^{-1}\}$  is bounded, where

$$P_j = (p_{j-1}, p_{j-2}, \dots, p_{j-n}).$$

Using this assumption and Lemma 3 we can now prove the following theorem.

#### Theorem 1

If Assumption 2 is satisfied then, for all update formulas (2.4) with

$$\beta_1, \beta_2 \geq 0, \beta_1 + \beta_2 \neq 0, \text{ we have}$$

$$i) \frac{\|x_{j+1} - z\|}{\|x_j - z\|} = o(\|x_{j-n} - z\|)$$

$$ii) \|H_j - G^{-1}\| = o(\|x_{j-n-1} - z\|).$$

#### Proof:

i) For  $j$  sufficiently large let

$$D_j = (d_{j-1}, d_{j-2}, \dots, d_{j-n}) \text{ and } P_j = P_j P_j.$$

Then

$$\begin{aligned} d_j &= G P_j P_j + (d_j - G P_j) \\ &= D_j P_j + (G P_j - D_j) P_j + (d_j - G P_j) \end{aligned}$$

and

$$\begin{aligned} M_j d_j &= P_j P_j + (M_j D_j - P_j) P_j + M_j (G P_j - D_j) P_j + M_j (d_j - G P_j) \\ &= P_j + (v_{j-1}, v_{j-2}, \dots, v_{j-n}, v_j) P_j + \tilde{v}_j, \end{aligned}$$

where

$$\begin{aligned} (3.7) \quad \|\tilde{v}_j\| &= o(\|H_j\| \|G P_j - D_j\| \|P_j^{-1}\| + \|d_j - G P_j\|) \\ &= o(\|g_{j-n}\|). \end{aligned}$$

Therefore,

$$M_j d_j = M_j (v_{j-1}, v_{j-2}, \dots, v_{j-n}, v_j) P_j + M_j \tilde{v}_j.$$

By Lemma 3 and (3.7) this equality implies

$$(3.8) \quad \|M_j d_j\| = o(\|g_{j-n}\|).$$

Using (3.8) and part i) of Lemma 2 we obtain

$$\frac{\|x_{j+1} - z\|}{\|x_j - z\|} \leq \frac{\gamma_j}{d_j^* P_j} \|M_j d_j\| + \frac{1}{d_j^* P_j} \frac{|q_{j+1}^* P_j|}{d_j^* P_j} = o(\|g_{j-n}\|),$$

which by Lemma 1 implies

$$\frac{\|x_{j+1} - z\|}{\|x_j - z\|} = o(\|x_{j-n} - z\|).$$

ii) It follows from Lemmas 1 and 3 and the first part of the theorem that, for  $i = j-2, \dots, j-n$ ,

$$\|v_{ij}\| = o(\|g_{j-1-n}\|).$$

Setting

$$V_j = (v_{j-1}, v_{j-2}, \dots, v_{j-n}, v_j)$$

we have

$$(M_j G - I) P_j = V_j - M_j (D_j - G P_j)$$

which implies

$$H_j - G^{-1} = V_j P_j^{-1} G^{-1} - H_j (D_j - G P_j) P_j^{-1} G^{-1}.$$

Therefore,

$$\begin{aligned} \|H_j - G^{-1}\| &= O(\|V_j\| + \|D_j - G P_j\|) \\ &= O(\|q_{j-n-1}\|), \end{aligned}$$

which by part v) of Lemma 1 completes the proof of the theorem.

For the special case  $\beta_1 = 1, \beta_2 = 0$ , i.e., for the Broyden-Fletcher-

Goldfarb-Shanno-method the above result has been obtained by Schuller [13].

For the following results we need a recurrence relation for the  $v_{ij}$ 's.

This will be derived in the following lemma.

Lemma 4

For every  $j \geq n$  and  $j-n \leq i < k < j$ ,

$$(3.9) \quad \|v_{i,k+1}\| = O\left(\frac{\|q_{k+1}\|}{\|q_k\|} \left(\frac{\|q_k' p_{i-1}\|}{\|q_k\|} + \|v_{ik}\| + \|q_i\|\right)\right).$$

Proof

By Taylor's theorem we have

$$(3.10) \quad q_{k+1} = q_k - \sigma_k G_k - \sigma_k E_k s_k,$$

where

$$E_k = \int_0^1 G(x_k - t(q_k s_k)) dt - G$$

and

$$\begin{aligned} (3.11) \quad \|E_k\| &\leq \max_{0 \leq t \leq 1} \|G(x_k - t(q_k s_k)) - G\| \\ &\leq \max_{0 \leq t \leq 1} \|L(x_k - t(x_k - x_{k+1})) - z\| \\ &\leq L \max(\|x_k - z\|, \|x_{k+1} - z\|) = O(\|q_k\|). \end{aligned}$$

Where the last relation follows from Lemma 1.

Multiplying (3.10) with  $P_i$  we obtain

$$\begin{aligned} (3.12) \quad q_{k+1}' P_i &= q_k' P_i - \sigma_k p_i' G_k - \sigma_k p_i' E_k s_k \\ &= q_k' P_i - \sigma_k d_i' q_k - \sigma_k (p_i' G - d_i') s_k - \sigma_k p_i' E_k s_k \\ &= (1 - \sigma_k) q_k' P_i - \sigma_k v_i' q_k - y_{ik}, \end{aligned}$$

where

$$y_{ik} = \sigma_k (p_i' G - d_i') s_k + \sigma_k p_i' E_k s_k$$

and because of (3.11) and Lemma 1

$$(3.13) \quad \|y_{ik}\| = O(\|q_k\| \|q_i\| + \|q_k\|^2) = O(\|q_k\| \|q_i\|).$$

Observing that by definition

$$p_i' d_k = \frac{p_i' q_k - p_i' q_{k+1}}{\|q_k\|}$$

we conclude from (3.12) and (3.13) that

$$\begin{aligned} (3.14) \quad |p_i' d_k| &= O\left(\frac{\|q_k' p_{i-1}\|}{\|q_k\|} + \|v_{ik}\| \frac{\|q_k\|}{\|s_k\|} + \frac{\|y_{ik}\|}{\|s_k\|}\right) \\ &= O\left(\frac{\|q_k' p_{i-1}\|}{\|q_k\|} + \|v_{ik}\| + \|q_i\|\right). \end{aligned}$$

Next assume that  $\|q_{k+1}\| \leq \|q_k\| \|q_i\|$ . Then it follows from Lemma 3 that

$$(3.15) \quad |v_k' d_i' q_k' p_i| \leq \eta^2 \|v_k\| = O\left(\frac{\|q_{k+1}\|}{\|q_k\|}\right) = O(\|q_i\|).$$

Now let

$$(3.16) \quad \|q_{k+1}\| > \|q_k\| \|q_i\|$$



and observe that by definition

$$(3.17) \quad d_{1,k}^* = d_{1,k}^* H_{1,k} = w_{1,k}^* p_1 + w_{1,k}^* v_{1,k},$$

where

$$(3.18) \quad w_k = \frac{q_{k+1}^{-1} \lambda_k q_k}{\|H_k(q_{k+1}^{-1} \lambda_k q_k)\|}, \quad \lambda_k = \frac{q_k^* p_k}{q_k^* p_k}.$$

Since  $w_k^* p_k = 0$ , it follows from Lemma 3 that

$$(3.19) \quad |w_k^* v_{1,k}| = O(\|q_1\|).$$

Using (3.16), (3.18) and Lemma 1 we obtain

$$\frac{\| \lambda_k q_k \|}{\| q_{k+1} \|} = \frac{\| q_{k+1}^* p_k \|}{\| q_{k+1} \|} \frac{\| q_k \|}{\| q_k^* p_k \|} = O\left(\frac{\| q_k \|^2}{\| q_{k+1} \|}\right) = O\left(\frac{\| q_k \|}{\| q_1 \|}\right).$$

Because  $\|q_k\|/\|q_1\| \rightarrow 0$  as  $j \rightarrow \infty$ , this relation implies that

$$\left\| H_k \left( \frac{q_{k+1}}{\|q_{k+1}\|} - \lambda_k \frac{q_k}{\|q_{k+1}\|} \right) \right\|$$

is bounded away from zero. By (3.18) we have, therefore, the equality

$$(3.20) \quad |w_k^* p_1| = O\left(\frac{\|q_{k+1}^* p_1\|}{\|q_{k+1}\|} + \|q_k\| \frac{\|q_k^* p_1\|}{\|q_{k+1}\|}\right).$$

Combining (3.12), (3.13) and (3.20) we see that

$$(3.21) \quad |w_k^* p_1| = O\left(\frac{\|q_k\|}{\|q_{k+1}\|} \left( \frac{\|q_k^* p_1\|}{\|q_k\|} + \|v_{1,k}\| + \|q_1\| \right) \right).$$

Observing that by Lemma 3,  $| \tau_k | = O(\|q_{k+1}\|/\|q_k\|)$  we deduce from (3.14), (3.15),

$$(3.17), (3.19) \text{ and } (3.21) \text{ the equality}$$

$$(3.22) \quad | \tau_k d_{1,k}^* q_k^* p_k | + | p_k^* d_k | = O\left(\frac{\|q_k^* p_1\|}{\|q_k\|} + \|v_{1,k}\| + \|q_1\| \right).$$

Since by Lemma 1,  $\|q_{k+1}\|/\|q_k\| = O(\|q_{k+1}\|/\|q_k\|)$ ; the desired result follows now from (3.22) and Lemma 3 with  $k = k+1$ .

Using the above lemma we can now generalize a result obtained by Stoer [14] who proved that, for all update formulas considered in this paper, the sequence  $\{\|x_j - z\|/\|x_{j-n} - z\|\}^2$  is bounded, provided  $x_0$  is sufficiently close to  $z$ . Without requiring that  $x_0$  is close to  $z$  we will prove the stronger result that the sequence  $\{\|x_j - z\|/\|x_{j-n} - z\|\}^2$  converges to zero.

#### Theorem 2

Let  $n \geq 2$ . Then, for every update formula (2.4) with  $\delta_1 \delta_2 \geq 0$ ,  $\delta_1 + \delta_2 \neq 0$ ,

$$\frac{\|q_j\|^2}{\|q_{j-n}\|^2} \rightarrow 0 \text{ as } j \rightarrow \infty$$

$$\frac{\|x_j - z\|}{\|x_{j-n} - z\|} \rightarrow 0 \text{ as } j \rightarrow \infty.$$

Proof:

Let  $j \geq n$  and  $j-n \leq i < k \leq j$ . We will first show that

$$(3.23) \quad \frac{|q_k^* p_1|}{\|q_k\|} = O\left(\frac{\|q_1\|^2}{\|q_k\|}\right)$$

and

$$(3.24) \quad \|v_{1,k}\| = O\left(\frac{\|q_1\|^2}{\|q_{k-1}\|}\right).$$

For  $k = i+1$  the two statements follow from Lemma 1 and  $v_{i,i+1} = 0$ , respectively. Now suppose that  $i+1 < k$  and (3.23) and (3.24) hold for some  $v$  with  $i < v < k$ . By (3.12) and (3.13)

$$\begin{aligned} \frac{|q_{v+1}^* p_1|}{\|q_{v+1}\|} &= O\left(\frac{\|q_v\|}{\|q_{v+1}\|} \left( \frac{|q_v^* p_1|}{\|q_v\|} + \|v_{1,v}\| + \|q_1\| \right) \right) \\ &= O\left(\frac{\|q_v\|}{\|q_{v+1}\|} \left( \frac{\|q_1\|^2}{\|q_v\|} + \frac{\|q_1\|^2}{\|q_{v-1}\|} + \|q_1\| \right) \right) \\ &= O\left(\frac{\|q_1\|^2}{\|q_{v+1}\|}\right). \end{aligned}$$

Similarly by (3.9),

$$\begin{aligned} \|v_{i,v+1}\| &= 0 \left( \frac{\|q_{v+1}\|}{\|q_v\|} \left( \frac{|q_v^i p_i|}{\|q_v\|} + \|v_{i,v}\| \right) + \|q_i\| \right) \\ &= 0 \left( \frac{\|q_i\|^2}{\|q_v\|} \left( \frac{\|q_{v+1}\|}{\|q_v\|} + \frac{\|q_{v+1}\|}{\|q_{v-1}\|} + \frac{\|q_v\|}{\|q_i\|} \right) \right) \\ &= 0 \left( \frac{\|q_i\|^2}{\|q_v\|} \right). \end{aligned}$$

This shows that (3.23) and (3.24) hold. Next we observe that

$$|p_k^i q_p| = |d_k^i p_i + (p_k^i - d_k^i) p_i| \leq |d_k^i p_i| + \|d_k - q_p\|.$$

Using Lemma 1, (3.14), (3.23) and (3.24) we obtain, therefore, the relation

$$(3.25) \quad |p_k^i q_p| = 0 \left( \frac{\|q_i\|^2}{\|q_k\|} \right).$$

Furthermore, it follows from (3.23) and (3.25) that, for  $i = j-n, \dots, k-1$  and

$$k = j-n+1, \dots, j-1,$$

$$(3.26) \quad |p_k^i q_p| = 0 \left( \frac{\|q_{j-n}\|^2}{\|q_{j-1}\|} \right) \quad \text{and} \quad \frac{|q_{j-1}^i p_i|}{\|q_{j-1}\|} = 0 \left( \frac{\|q_{j-n}\|^2}{\|q_{j-1}\|} \right).$$

To complete the proof we assume now that there are  $\epsilon > 0$  and an infinite

subset  $J \subset \{0, 1, \dots\}$  such that

$$(3.27) \quad \frac{\|q_j\|}{\|q_{j-n}\|} \geq \epsilon \quad \text{for } j \in J.$$

Since  $\|q_j\|/\|q_{j-1}\| \rightarrow 0$  as  $j \rightarrow \infty$  this implies that

$$\frac{\|q_{j-n}\|^2}{\|q_{j-1}\|} \rightarrow 0 \quad \text{as } j \rightarrow \infty, j \in J.$$

Hence it follows from (3.26) that there are a  $j_0$  and a constant  $\delta$  such that for  $j \geq j_0, j \in J$  the matrix

$$P_j = (p_{j-1}, p_{j-2}, \dots, p_{j-n})$$

is nonsingular and

$$(3.28) \quad \|P_j^{-1}\| \leq \delta.$$

Using Lemma 1, (3.12), (3.13), (3.24) and (3.26) we conclude that

$$|q_j^i p_{j-1}| = 0 \left( \|q_{j-1}\|^2 \right) = 0 \left( \frac{\|q_{j-1}\|^2}{\|q_{j-n}\|^2} \|q_{j-n}\|^2 \right)$$

and for  $i = j-2, \dots, j-n$ ,

$$|q_j^i p_i| = 0 \left( |1 - \sigma_{j-1}| \|q_{j-n}\|^2 + \frac{\|q_{j-n}\|^2}{\|q_{j-2}\|} \|q_{j-1}\| + \|q_{j-1}\| \|q_{j-n}\| \right).$$

Since by Lemma 1,  $|1 - \sigma_j| \rightarrow 0$  as  $j \rightarrow \infty$  this implies that, for  $i = j-1, \dots, j-n$

$$\frac{|q_j^i p_i|}{\|q_{j-n}\|^2} \rightarrow 0 \quad \text{as } j \rightarrow \infty, j \in J.$$

In conjunction with (3.28) this shows that

$$\frac{\|q_j\|}{\|q_{j-n}\|} \rightarrow 0 \quad \text{as } j \rightarrow \infty, j \in J.$$

Because this is a contradiction to (3.27) it follows that

$$\|q_j\|/\|q_{j-n}\|^2 \rightarrow 0 \quad \text{as } j \rightarrow \infty,$$

which in view of Lemma 1 completes the proof of the theorem.

As we have seen in Section 2 the matrix  $H_j$  can be represented in the form

$$(3.29) \quad H_j = \frac{p_j p_j^i}{p_j q_j^i p_j} + \sum_{i=2}^n \frac{p_i p_i^i}{d_i q_j^i p_j}$$

where  $\|P_j\| = \|P_{2j}\| = \dots = \|P_{nj}\| = 1$ . It is shown in the next lemma that an estimate for the rate of superlinear convergence can be obtained by using the numbers  $\|d_{ij} - Gp_{ij}\|$ ,  $i = 2, \dots, n$ .

#### Lemma 5

Let  $H_j$  be given by (3.29). Then

$$i) \frac{\|g_{j+1}'\|}{\|g_j\|} = O(\|g_j\|)$$

$$\frac{\|g_{j+1}'\|}{\|g_j\|} = O(\|d_{ij} - Gp_{ij}\| + \|g_j\|), i = 2, \dots, n$$

$$ii) \frac{\|g_{j+1}'\|}{\|g_j\|} = O\left(\max \left\{ \frac{\|g_{j+1}'\|}{\|g_j\|}, \frac{\|g_{j+1}'\|}{\|g_j\|}, i = 2, \dots, n \right\}\right).$$

#### Proof.

Replacing  $x$  with  $j$  and multiplying with  $P_{ij}$  we obtain from (3.10) the equality

$$\begin{aligned} g_{j+1}'P_{ij} &= g_j'P_{ij} - \sigma_j P_{ij}' G s_j - \sigma_j P_{ij}' E s_j \\ &= -\sigma_j d_{ij}' H_j s_j - \sigma_j (P_{ij}' G - d_{ij}') s_j - \sigma_j P_{ij}' E s_j \\ &= -\sigma_j (P_{ij}' G - d_{ij}') s_j - \sigma_j P_{ij}' E s_j, \end{aligned}$$

which by (3.11) implies

$$\frac{\|g_{j+1}'\|}{\|g_j\|} = O\left(\frac{\|s_j\|}{\|g_j\|} (\|d_{ij} - Gp_{ij}\| + \|g_j\|) + \|g_j\|\right).$$

In connection with part iii) of Lemma 1 this completes the proof of the first part of the lemma.

Let  $x \in E^n$  and

$$(3.30) \quad x = \lambda_0 g_j + \sum_{i=2}^n \lambda_i d_{ij}.$$

Since it follows from part v) of Assumption 1 that  $\|H_j^{-1}\| \leq 1/\tau_1$  we have

$$\|x\| = O(\max\{|\lambda_1|, |\lambda_2|, \dots, |\lambda_n|\}).$$

Multiplying both sides of (3.30) with  $P_j$  and  $P_{ij}$ ,  $i = 2, \dots, n$ , gives

$$P_j' x = \lambda_0 g_j' P_j \text{ and } P_{ij}' x = \lambda_i d_{ij}' P_{ij}.$$

Because  $\rho_j g_j' P_j = P_j' H_j^{-1} P_j \geq 1/\tau_2$  and  $d_{ij}' P_{ij} = P_{ij}' H_j^{-1} P_{ij} \geq 1/\tau_2$  this completes the proof of the lemma.

According to the above lemma we have

$$(3.31) \quad \frac{\|g_{j+1}'\|}{\|g_j\|} = O(\max\{\|d_{ij} - Gp_{ij}\|, i = 2, \dots, n\} + \|g_j\|).$$

It is interesting to observe that this relation is independent of the first term on the right hand side of (3.29), i.e., of  $\|p_j g_j - Gp_j\|$ .

It follows from (2.6) and part iv) of Lemma 1 that there is a representation of  $H_{j+1}$  in terms of  $n$  matrices of rank one containing the term

$$\frac{P_j P_j'}{d_j' P_j} \text{ with } \|d_j - Gp_j\| = O(\|g_j\|).$$

Similarly, an analogous representation of  $H_{j+2}$  contains a term

$$\frac{P_{j+1} P_{j+1}'}{d_{j+1}' P_{j+1}} \text{ with } \|d_{j+1} - Gp_{j+1}\| = O(\|g_{j+1}\|).$$

This observation suggests that, under certain assumptions, it might be possible to prove that

$$(3.32) \quad \max\{\|d_{ij} - Gp_{ij}\|, i = 2, \dots, n\} = O(\|g_{j-n+1}\|)$$

which by (3.11) would imply

$$(3.32) \quad \frac{\|g_{j+1}\|}{\|g_j\|} = O(\|g_{j-n+1}\|).$$

Such a result can indeed be obtained if we assume that a certain lower bound on the rate of convergence, as specified in the following assumption, is valid.

#### Assumption 2

Let  $H_j$  be given by (3.29) and assume that there is  $\delta > 0$  such that for every  $j$

$$\frac{\|g_{j+1}\|}{\|g_j\|} \geq \delta \max_{1 \leq i \leq n} (\|d_{1j} - Gp_{1j}\|, \|g_{j-n+1}\|).$$

In view of (3.31) and the discussion leading to (3.32) Assumption 2 implies that the sequence  $\{\|g_{j+1}\|/\|g_j\|\}$  does not converge faster than could be expected at best under the given circumstances for a general function  $F(x)$ .

As a first step towards establishing (3.33) we prove the following lemma.

#### Lemma 5

Let Assumption 2 be satisfied and let

$$H_j = \frac{P_j P_j^T}{\rho_j q_j^T P_j} + \sum_{i=2}^n \frac{P_{1j} P_{1j}^T}{d_{1j}^T P_{1j}}.$$

Then there are a constant  $\tau > 0$ , independent of  $j$ , and vectors  $d_{1,j+1}$ ,  $P_{1,j+1}$ ,  $i = 2, \dots, n$ , such that

$$i) \quad H_{j+1} d_{1,j+1} = P_{1,j+1}, \quad \|P_{1,j+1}\| = 1, \quad i = 2, \dots, n$$

$$ii) \quad H_{j+1} = \frac{P_{1,j+1} P_{1,j+1}^T}{\rho_{j+1} q_{j+1}^T P_{j+1}} + \sum_{i=2}^n \frac{P_{1,j+1} P_{1,j+1}^T}{d_{1,j+1}^T P_{1,j+1}}$$

$$iii) \quad \|d_{j+2} g_{j+1} - Gp_{j+2}\| \leq \tau \max(\|g_j\|, \|g_{j+1}\|/\|g_j\|, \|d_{1j} - Gp_{1j}\|, i = 2, \dots, n)$$

$$iv) \quad \|d_{2,j+1} - Gp_{2,j+1}\| \leq \tau \|g_j\|$$

$$v) \quad \|d_{1,j+1} - Gp_{1,j+1}\| \leq \tau \max(\|d_{1-1,j} - Gp_{1-1,j}\|, \|g_j\|), \quad i = 3, \dots, n.$$

Proof:

Let

$$q_j = \frac{g_{j+1} - \lambda g_j}{\|H_j(g_{j+1} - \lambda g_j)\|}, \quad q_j = \frac{H_j(g_{j+1} - \lambda g_j)}{\|H_j(g_{j+1} - \lambda g_j)\|}, \quad \lambda_j = \frac{g_{j+1}^T P_j}{g_j^T P_j}.$$

Then  $w_j^T P_j = 0$  and

$$w_j \in \text{span}(d_{2j}, \dots, d_{nj}).$$

By part v) of Assumption 1 this implies that

$$(3.34) \quad \|w_j - Gq_j\| = O(\max(\|d_{1j} - Gp_{1j}\|, i = 2, \dots, n)).$$

Define

$$\tilde{P}_{1j} = \frac{P_j}{\rho_j q_j^T P_j}, \quad \tilde{P}_{1j} = \frac{P_{1j}}{d_{1j}^T P_{1j}}, \quad i = 2, \dots, n$$

and set

$$P_j = (\tilde{P}_{1j}, \dots, \tilde{P}_{nj}), \quad v_j = w_j^T P_j,$$

$$(3.35) \quad z_j = \frac{c_j \|v_j\| e_{-v_j}}{\|c_j\| \|v_j\| e_{-v_j}}, \quad z_j = (z_1, \dots, z_n),$$

where

$$e_n = (0, \dots, 0, 1) \quad \text{and} \quad c_j = \begin{cases} 1 & \text{if } (v_j)_n \leq 0 \\ -1 & \text{if } (v_j)_n > 0. \end{cases}$$

Then

$$Q_j = I - 2z_j z_j^T$$



is a Householder - matrix with the property (see [10] for instance)

$$(3.36) \quad Q_j^* v_j = e_j \|v_j\| e_n.$$

Let

$$(3.37) \quad (\hat{q}_n, \dots, \hat{q}_3) = P_j Q_j = (\hat{p}_1, \dots, \hat{p}_n - 2\hat{c}_1 \hat{p}_1, \dots, \hat{p}_n - 2\hat{c}_n \hat{p}_n).$$

Since  $H_j = P_j P_j^*$ , we have

$$(3.38) \quad P_j Q_j (P_j Q_j)^* = P_j (Q_j Q_j^*) P_j^* = P_j P_j^* = H_j.$$

Furthermore, it follows from  $w_j^* p_j = 0$  and (3.36), respectively, that

$$(3.39) \quad \hat{q}_1 = \hat{p}_1 \quad \text{and} \quad \hat{q}_n = q_n \|\hat{q}_n\|.$$

Defining

$$(3.40) \quad p_{i,j+1} = \hat{q}_{i-1,j} / \|\hat{q}_{i-1,j}\|, \quad d_{i,j+1} = H_j^{-1} p_{i,j+1}, \quad i = 3, \dots, n$$

we deduce from (3.37) through (3.40) that

$$H_j = \frac{p_j p_j^*}{p_j^* p_j} + \frac{q_j q_j^*}{q_j^* q_j} + \sum_{i=2}^n \frac{p_{i,j+1} p_{i,j+1}^*}{d_{i,j+1}^* d_{i,j+1}}.$$

By (3.37) and (3.40) we have

$$(3.41) \quad p_{i,j+1} = \frac{1}{\|\hat{q}_{i-1,j}\|} (\hat{p}_{i-1,j} - 2\hat{c}_{i-1} \hat{p}_{i-1,j} - \sum_{v=1}^n \hat{c}_v \hat{p}_{v,j}), \quad i = 3, \dots, n$$

$$(3.42) \quad d_{i,j+1} = \frac{1}{\|\hat{q}_{i-1,j}\|} (H_j^{-1} \hat{p}_{i-1,j} - 2\hat{c}_{i-1} H_j^{-1} \hat{p}_{i-1,j} - \sum_{v=1}^n \hat{c}_v H_j^{-1} \hat{p}_{v,j}), \quad i = 3, \dots, n.$$

Since  $c_1 = 0$ ,  $\|c_v\| \leq 1$ ,  $v = 2, \dots, n$ ,  $\hat{q}_1^* H_j^{-1} \hat{q}_1 = 1$  and

$$\|H_j^{-1} \hat{p}_{1,j} - \hat{c} \hat{p}_{1,j}\| = 0 (\|d_{1,j} - \hat{c} p_{1,j}\|), \quad i = 2, \dots, n,$$

it follows from (3.41) and (3.42) that, for  $i = 3, \dots, n$ ,

$$(3.43) \quad \|d_{i,j+1} - \hat{c} p_{i,j+1}\| = 0 (\|d_{i-1,j} - \hat{c} p_{i-1,j}\| + \|c_{i-1}\| \sum_{v=2}^n \|d_{v,j} - \hat{c} p_{v,j}\|).$$

Observing that

$$\|e_j \|v_j \|e_n - v_j\|^2 \geq \|v_j\|^2 = w_j^* p_j p_j^* w_j = w_j^* q_j$$

is bounded away from zero we deduce from (3.35) the equality

$$(3.44) \quad |c_i| = 0 (|v_j|) = 0 (w_j^* p_{1,j}) = 0 (w_j^* p_{1,j}), \quad i = 2, \dots, n-1.$$

Since by assumption  $\|q_{j+1}\| \geq \delta \|q_j\| \|q_{j-n+1}\|$  it follows from Lemma 1 that

$$\lambda_j = q_{j+1}^* p_j / q_j^* p_j + 0 \text{ as } j \rightarrow \infty$$

which as in the proof of Lemma 4 implies that

$$\|H_j \left( \frac{q_{j+1}}{\|q_{j+1}\|} - \lambda_j \frac{q_j}{\|q_{j+1}\|} \right)\|$$

is bounded away from zero. Thus we conclude from (3.44),  $q_j^* p_{1,j} = 0$ , and

Lemma 5 that, for  $i = 2, \dots, n-1$ ,

$$(3.45) \quad |c_i| = 0 \left( \frac{q_{i+1}^* p_{1,j}}{\|q_{i+1}\|} \right) = 0 \left( \frac{\|d_{i,j} - \hat{c} p_{i,j}\| + \|q_i\|}{\max\{\|d_{1,j} - \hat{c} p_{1,j}\|, i=2, \dots, n\}} \right).$$

Therefore, part v) of the lemma follows from (3.43) and (3.45).

By (2.6)

$$H_{j+1} = \frac{p_{j+1} p_{j+1}^*}{d_{j+1}^* d_{j+1}} + w_j \frac{v_j v_j^*}{w_j^* w_j} + \sum_{i=3}^n \frac{p_{i,j+1} p_{i,j+1}^*}{d_{i,j+1}^* d_{i,j+1}},$$

where

$$v_j = \frac{q_{j+1} p_j}{\|q_{j+1} p_j\|}, \quad a_j = -\frac{d_{1,j}^* q_1}{d_{j+1}^* p_j}.$$

Since by Lemma 1 and (2.7),  $|g_j| = o(\|g_{j+1}\|/\|g_j\|)$ ,  $|w_{j-1}| = o(\|g_{j+1}\|/\|g_j\|)$

it follows that

$$(3.46) \quad \left\| \frac{w_j}{w_j} - Gg_j \right\| = o(\|w_j - Gg_j\| + \|g_{j+1}\|/\|g_j\|).$$

Furthermore,  $g_{j+1} \in \text{span}(d_{j+1}, w_j)$  implies that

$$(3.47) \quad p_{j+1}g_{j+1} = \lambda_j d_j + \tau_j \frac{w_j}{w_j}, \quad p_{j+1} = \lambda_j p_j + \tau_j w_j.$$

Observing that by Lemma 1,  $\|g_j - Gp_j\| = o(\|g_j\|)$  and  $|\lambda_j| \leq \bar{\lambda}$ ,  $|\tau_j| \leq \bar{\tau}$  for some  $\bar{\lambda}$  and  $\bar{\tau}$ , independent of  $j$ , we obtain from (3.46) and (3.47) the relation

$$(3.48) \quad \|p_{j+1}g_{j+1} - Gp_{j+1}\| = o(\|w_j - Gg_j\| + \|g_j\| + \|g_{j+1}\|/\|g_j\|).$$

Pirally, let

$$(3.49) \quad p_{2,j+1} = \bar{\lambda}_j p_j + \bar{\tau}_j p_{j+1}$$

be such that  $\|p_{1,j+1}\| = 1$  and  $g_{j+1}p_{2,j+1} = 0$ . Then

$$(3.50) \quad |\bar{\tau}_j| = |\bar{\lambda}_j| \frac{|g_{j+1}p_j|}{g_{j+1}p_{j+1}} = o\left(|\bar{\lambda}_j| \frac{\|g_{j+1}\|}{g_{j+1}p_{j+1}} \frac{\|g_j\|^2}{\|g_{j+1}\|}\right) = o\left(\frac{\|g_j\|^2}{\|g_{j+1}\|}\right),$$

where the last two equalities follow from Lemma 1 and the fact that  $\{\bar{\lambda}_j\}$  is bounded. Therefore, defining

$$d_{2,j+1} = h_{j+1}^{-1} p_{2,j+1}$$

we obtain from (3.48), (3.49) and (3.50)

$$(3.51) \quad \|d_{2,j+1} - Gp_{2,j+1}\| \leq |\bar{\lambda}_j| \|d_j - Gp_j\| + |\bar{\tau}_j| \|p_{j+1}g_{j+1} - Gp_{j+1}\| \\ = o(\|g_j\| + |\bar{\tau}_j| (\|w_j - Gg_j\| + \|g_{j+1}\|/\|g_j\|)) \\ = o(\|g_j\| + \frac{\|g_j\|}{\delta v_j} (\|w_j - Gg_j\| + \|g_j\|)) \\ = o(\|g_j\|).$$

where

$$v_j = \max(\|g_{j-n+1}\|, \|d_{1j} - Gp_{1j}\|, i = 2, \dots, n)$$

and the last equality follows from (3.34) and Assumption 3.

Since  $g_{j+1}p_{2,j+1} = 0$  we can now represent  $H_{j+1}$  in the form

$$H_{j+1} = \frac{p_{j+1}p_{j+1}'}{p_{j+1}g_{j+1}p_{j+1}} + \sum_{i=2}^n \frac{p_{i,j+1}p_{i,j+1}'}{d_{i,j+1}p_{i,j+1}}.$$

In conjunction with (3.48), and (3.51) this completes the proof of the lemma.

A repeated application of Lemma 6 shows that the estimate (3.32) is valid and leads to the following theorem.

### Theorem 3

Let Assumption 1 and 3 be satisfied. Then for every update formula (2.4)

with  $\theta_1 \theta_2 \geq 0$ ,  $\theta_1 + \theta_2 \neq 0$  the following statements hold.

- i)  $\frac{\|x_{j+1} - z\|}{\|x_j - z\|} = o(\|x_{j-n+1} - z\|)$
- ii)  $\|H_j - G^{-1}\| = o(\|x_{j-n} - z\|)$
- iii)  $\|1 - \sigma_j\| = o(\|x_{j-n} - z\|)$ .

Proof.

If we write each  $H_j$  in the form

$$H_j = \frac{p_{j1} p_{j1}'}{p_{j2} p_{j2}'} + \sum_{i=2}^n \frac{p_{ji} p_{ji}'}{d_{ji} p_{ji}'}$$

it follows from Lemma 6 that

$$(3.52) \quad \|d_{1j} - Gp_{1j}\| = O(\|g_{j-1}\|), \quad j = 2, \dots, n,$$

which by Lemma 5 implies

$$(3.53) \quad \frac{\|g_{j+1}\|}{\|g_j\|} = O(\|g_{j-n+1}\|).$$

The first statement of the theorem follows now from (3.53) and part v) of

Lemma 1. Furthermore, we obtain from (3.52), (3.53) and part iii) of Lemma 6 the relation

$$(3.54) \quad \|p_{jj} - Gp_j\| = O(\|g_{j-n}\|).$$

Observing that, by Lemma 1, the sequence

$$\{(p_{j1}, p_{j2}, \dots, p_{jn})^{-1}\}, \quad j = 1, 2, \dots$$

exists and is bounded we deduce from (3.52) and (3.54) that

$$\|H_j - G^{-1}\| = O(\|g_{j-n}\|) = O(\|x_{j-n} - z\|).$$

Finally, it follows from Taylor's theorem that there is

$$Y_j \in [x \mid x = x_j - ts_j, \quad 0 \leq t \leq 1]$$

such that

$$2[F(x_j - s_j) - F(x_j) + g_j s_j] = s_j' G(Y_j) s_j.$$

Because

$$s_j' G(Y_j) s_j = g_j' s_j + s_j' G(H_j - G^{-1}) g_j + s_j' (G(Y_j) - G) s_j$$

we obtain from the definition of  $g_j$  (see (2.9)) the relation

$$\begin{aligned} |1 - \alpha_j| &= \frac{|s_j' G(Y_j) s_j - g_j' s_j|}{s_j' G(Y_j) s_j} \\ &= O\left(\frac{\|g_j\|}{\|s_j\|}\right) \|G\| \|H_j - G^{-1}\| + \|G(Y_j) - G\| \\ &= O(\|x_{j-n} - z\|), \end{aligned}$$

where the last equality follows from

$$\|G(Y_j) - G\| \leq 2 \max(\|x_j - z\|, \|x_j - s_j - z\|) = O(\|x_j - z\|).$$

As a further consequence of Assumption 3 we have the following theorem which implies that  $n$  consecutive search directions are asymptotically conjugate with respect to the Hessian matrix of  $F(x)$  at  $z$ .

#### Theorem 4

Let Assumptions 1 and 3 be satisfied. Then for every update formula (2.4)

with  $\beta_1 \beta_2 \geq 0$ ,  $\beta_1 + \beta_2 \neq 0$  the following statements hold.

$$i) \quad \|H_j d_j - p_j\| = O(\|g_j\|), \quad j = j-1, \dots, j-n$$

$$ii) \quad \frac{|g_k' p_j|}{\|g_k\|} = O\left(\frac{\|g_j\|}{\|g_{k-n}\|}\right), \quad j-n \leq k < j$$

$$iii) \quad |p_k' G p_j| = O\left(\frac{\|g_j\|}{\|g_{k-n}\|}\right), \quad j-n \leq k < j.$$

Proof.

Let  $j \geq n$  and  $j-n \leq k < j$ . We will first show that

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$$(3.55) \quad \frac{|q_k' p_1|}{\|q_k\|} = O\left(\frac{\|q_1\|}{\|q_{k-n}\|}\right) \quad \text{and} \quad \|v_{1k}\| = O(\|q_1\|).$$

For  $k = i+1$  the two statements follow from Lemma 1 and Lemma 3, respectively. Suppose that  $i+1 < k$  and (3.55) holds for some  $i < v < k$ . By (3.12), (3.13), and Theorem 3

$$\begin{aligned} \frac{|q_{v+1}' p_1|}{\|q_{v+1}\|} &= O\left(\frac{\|q_1\|}{\|q_{v-n}\|} \frac{\|q_v' p_1\|}{\|q_v\|} + \frac{\|q_v\|}{\|q_{v+1}\|} \left(\|v_{1v}\| + \|q_1\|\right)\right) \\ &= O\left(\frac{\|q_{v-n}\|}{\|q_{v-n+1}\|} \frac{\|q_1\|}{\|q_{v-n}\|} + \frac{\|q_1\|}{\|q_{v-n+1}\|}\right) \\ &= O\left(\frac{\|q_1\|}{\|q_{v-n+1}\|}\right). \end{aligned}$$

Similarly by (3.9) and Theorem 3

$$\begin{aligned} \|v_{1,v+1}\| &= O\left(\frac{\|q_{v+1}\|}{\|q_v\|} \left(\frac{|q_v' p_1|}{\|q_v\|} + \|v_{1v}\|\right) + \|q_1\|\right) \\ &= O\left(\frac{\|q_{v-n+1}\|}{\|q_{v-n}\|} \left(\frac{\|q_1\|}{\|q_{v-n}\|} + \|q_1\|\right) + \|q_1\|\right) \\ &= O(\|q_1\|). \end{aligned}$$

Since  $v_{1j} = H_j d_j - p_1$  this completes the proof of the first two parts of the theorem. Finally we observe that

$$|p_k' G p_1| = |d_k' p_1 + (p_k' G - d_k') p_1| \leq |d_k' p_1| + \|d_k - G p_k\|,$$

which by (3.14), Lemma 1 and the first two parts of the theorem implies

$$|p_k' G p_1| = O\left(\frac{|q_k' p_1|}{\|q_k\|} + \|v_{1k}\| + \|q_1\|\right) = O\left(\frac{\|q_1\|}{\|q_{k-n}\|}\right).$$



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